

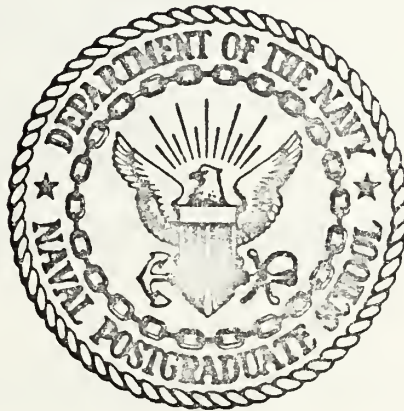
AN ANALYSIS OF THE NAVAL PERSONNEL
PAY PREDICTOR (ENLISTED MODEL)

Allan Ray Walker

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THESIS

AN ANALYSIS OF THE
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(ENLISTED MODEL)

by

Allan Ray Walker

September 1975

Thesis Advisor:

R.W. Butterworth

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(20. ABSTRACT Continued)

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An Analysis of the
Naval Personnel Pay Predictor
(Enlisted Model)

by

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ABSTRACT

The Naval Personnel Pay Predictor (enlisted Model) is used by the Bureau of Naval Personnel as a tool for predicting the total annual basic pay for the enlisted force as an input to the budget process. A major source of error in the model was found to be the prediction of the length of service (LOS) vector, and an attempt to improve this prediction was made. The extreme complexity of the model was found to be unnecessary, and a simple exponential smoothing subroutine for LOS prediction did as well or better than the original model. It was also found that a double exponential smoothing subroutine, taking into account the trends in the force structure, would almost uniformly improve the one year prediction from the model.

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I. NATURE OF THE PROBLEM

The Bureau of Naval Personnel requires many mathematical models for accurately predicting the structure of the future force. These models are used as tools to aid in planning decisions. Of special interest is the problem of costing the future force as a part of the budget submission procedure.

The Bureau of Naval Operations determines the personnel requirements for the future force and passes these to the Bureau of Personnel for implementation. These requirements are presented in the form of quarterly pay grade vectors, that is, the number of people required in pay grade E-1, E-2, E-3, ..., E-9. Since the amount of pay received is dependent on the member's length of service, the problem of predicting the total cost of the force becomes complex.

The specific problem considered was: given the future size of the force by pay grade, the past and present inventories, predict the total annual base pay of the force for future years.

II. THE NAPPE MODEL

One model currently used for this purpose is the Naval Personnel Pay Predictor (Enlisted Model), referred to as NAPPE. The model makes use of a data base consisting of three sets of quarterly inventories (pay grade by LOS Force matrices) for all years since 1957. The inventories are for United States Navy (USN), United States Naval Reserve (USNR), and Total Navy (TOTALNAV), the sum of the two.

The procedure is to, first, predict the future quarterly LOS vectors for the desired number of years into the future (up to ten). This is a vector of the total number of people with length of service 1, 2, ..., 31 years. The methodology used for this prediction is discussed later in this section and in Appendices A and B. The LOS vector is then combined with the pay grade requirements vector to get the predicted force matrix. A discussion of this procedure is also included in this section. The cost of the force is then simply the multiplication of the straight line averaged (between successive quarters) number of people in each cell of the matrix with the pay scale for that cell, which is an input to the model.

A. SMOOTHING THE LOS VECTORS OF THE INVENTORIES

The first step in the NAPPE models prediction of the LOS vector is accomplished by a subroutine referred to as SMOOTH

(refer to Appendix A for the mathematical model). Throughout the discussion of SMOOTH it should be remembered that all calculations applied to previous years data are made independently for the population (the total number of people) in each element of the LOS vector (hence referred to as LOS cell) and the transition rates from one cell to the next, computed for all three data bases. The transition rate is simply the proportion of the population in cell i of year j which move to cell $i+1$ in year $j+1$. The methodology is basic single exponential smoothing as discussed by Brown [1] and others.

The following procedure is done independently for each LOS cell. For each year of historical data, a prediction is made based on exponentially smoothing the data up to that year using values of the smoothing constant (hence referred to as α) of .05, .10, ..., .95. For each year, the predicted value is then compared with the actual value to determine which value of α would have given the best prediction. This results in the selection of an α for each LOS cell for each year of data. Consult Appendix A for the exact procedure and forms of the resulting error that are stored and used by the model. "Best" predictions and resulting errors are made for all years of historical data, finally resulting in a decision for the "best" α for predicting the future.

The output of SMOOTH consists of four sets of LOS vectors. For each year of historical data, there is a prediction based

on transition rates and a prediction based on previous year's cell populations, one pair based on the TOTALNAV data and the other pair on the sum of the predictions based on USN and USNR data. Note that due to the difference in the structure of the USN and USNR, the sum of these predictions may be different than the prediction based on their sum.

The final prediction is made by a subroutine referred to as ADJSMO (refer to Appendix B for the model). ADJSMO considers five "methods" for prediction. These include the four outputs from SMOOTH plus a weighted average of these predictions. This weighted average is formed by multiplying a weight (BWT) times the average of the two transition rate based predictions plus the complementary weight ($1.0 - \text{BWT}$) times the average of the two population based predictions. This calculation is made for values of BWT of .45, .50, ..., .95. For each year of data a "best" method (of the five), in the least square error sense, and a "best" weight, if a weighted average method was chosen, is selected for predicting that year. The absolute sum of the errors of the "best" prediction is also calculated for use in adjusting the final prediction. This adjustment is necessary because no transition rate is available to predict LOS cell 1.

Having selected a "best" method and a "best" weighting factor based on the last year of historical data, the model predicts the first future year values for LOS cells 2-31. At this point the model calculates the average (over all

years of data) proportion of the total population that was in LOS cell 1. This proportion is then applied to the total force required for the quarter under consideration and compared with the number which would be in cell 1 given the predicted values for cells 2-31 and the required total. Half of the difference in these two values is then allocated among cells 2-31 according to the total absolute error discussed above. The prediction for cell 1 is then the difference between the required total size of the force and the predictions made for cells 2-31.

B. GENERATING THE PAY GRADE BY LOS MATRIX

The pay grade by LOS matrix is calculated using a method for renormalizing contingency tables, as described by Mosteller [Ref. 3]. This method is an iterative procedure which takes the desired marginal totals of a matrix and a given, or base, matrix of the desired form and constructs a matrix as similar as possible to the base matrix, having the marginal totals that were desired.

Since this method was used throughout the research, a brief discussion of the procedure follows:

Let $A_{i,j}$ be the elements of the base matrix, $i = 1, 2, \dots, 9$
 $j = 1, 2, \dots, 31$

Let R_j be the desired row totals

Let C_i be the desired column totals

$$(1) \quad R_j^! = \sum_{i=1}^9 A_{i,j} \quad \text{for all } j$$

$$D_j = R_j/R'_j \quad \text{for all } j$$

$$A'_{i,j} = D_j A_{i,j} \quad \text{for all } i,j$$

$$C'_i = \sum_{j=1}^{31} A'_{i,j} \quad \text{for all } i$$

$$D'_i = C_i/C'_i \quad \text{for all } i$$

$$A''_{i,j} = D'_i A'_{i,j} \quad \text{for all } i,j$$

$$A_{i,j} = A''_{i,j}$$

Return to step (1).

The procedure is continued until the row and column totals converge to the desired totals.

In the NAPPE model, the marginals are the given pay grade vector and the predicted LOS vector. The base matrix is calculated as the simple average of the last twelve quarterly historic inventories.

III. EXPERIMENTS WITH AND CHANGES TO THE NAPPE MODEL

Since the object of the research was to improve the predictive accuracy of the NAPPE model, the sources of error had to be determined. It appeared that there were two independent sources of error, predicting the LOS vector and the instability of the Mosteller procedure for completing the matrix. Both of these possible problem areas were studied. In this section is a discussion of the first area and changes which were made to the model to improve its predictive quality. This is followed by a discussion of a study to discover the factors which influence the Mosteller procedure.

A. MAJOR SOURCE OF ERROR

The removal of either of these above mentioned sources of error should improve the predictions of the model. An ad hoc test of this hypothesis was accomplished by using the Mosteller subroutine (called PNGPNG) with the correct LOS vector and comparing the results of the model with known values, for years with historical data available.

The NAPPE model has a validation feature which facilitates this and other kinds of comparisons. As an input to the model, the last date of historical data to be used is given. The model then only looks at data up to that date and predicts as if that were today's date. Also included in the NAPPE package (which consists of several minor models besides

NAPPE itself) is a model called NAPVAL. This model compares the NAPPE output with the actual inventories. These comparisons are discussed throughout this paper. Specifically, any number called "actual" will mean an output from NAPVAL. Also, throughout the paper, the measure of effectiveness for comparison will be the total annual cost of the force, which is an output of both models.

In order to accomplish the above mentioned objective, the SMOOTH and ADJSMC subroutines were removed from the model. In their place, the actual LOS vector was read from the inventories and the following table is the result of comparing the prediction based on this procedure and the prediction of NAPPE. The elements of the table are the actual cost (NAPVAL), the NAPPE prediction (with the model untouched), and the prediction using only the Mosteller procedure (NAPPE with SMOOTH and ADJSMO removed), labeled "Using Actual LOS".

TABLE I

COMPARISON OF NAPPE WITH PURE MOSTELLER
(costs are in millions of dollars)

Year	Actual Cost	NAPPE Prediction	% Error	Using Actual Los	% Error
1968	1,832	1,839	.363	1,832	.014
1969	2,009	2,016	.372	2,010	.061
1970	2,280	2,286	.262	2,282	.081
1971	2,264	2,258	.290	2,266	.092

These results indicated that a major source of error in the model, and hence a potential for improvement, results from the prediction of the LOS vector, as was expected. If one looks at any feature of the enlisted force (such as size or distribution), he finds that it is not stationary in time, even considering statistical fluctuations. There are obvious trends. During war years, the force becomes larger and, on the average, younger, while during peace time, the force becomes smaller and older. Since single exponential smoothing does not allow for trends, it could not be expected to handle the problem being considered.

However, before attacking this problem, there were other questions to be answered. After documenting the model, two other questions came to mind. Is the pure complexity of the model worth the computer requirements? (The following section indicates not.) Is the use of the entire data base justified? Intuitively, the answer to the second question was no. The size and structure of the force in the late 1950's is not indicative of the force in 1975. There are continual policy changes which affect enlistment, promotion, and retention.

B. SIMPLE EXPONENTIAL SMOOTHING

In order to answer these questions, the first change in the model was made. The SMOOTH and ADJSMO subroutines were removed and a subroutine, SMOTHY, replaced them. This

subroutine used simple exponential smoothing of the transition rates and only the TOTALNAV data base.

1. The SMOTHY Subroutine

This subroutine results in an extensive simplification of the NAPPE model as it uses only four years of historical data and a single alpha value of 0.4. This alpha value may seem very large, but it was desired to make the prediction extremely dependent on the most recent data, which is the most significant. The actual subroutine is included at the end of the paper but the simple mathematical model follows:

Let $A_{i,j,k}$ be the number of people in LOS cell k in quarter j of year i

For the four years of historical data calculate the loss rate for each quarter and each LOS independently

$$TR_{i,j,k} = \frac{A_{i,j,k} - A_{i+1,j,k+1}}{A_{i,j,k}}$$

The following series of data were smoothed for prediction

..., $TR_{i,1,k}$, $TR_{i,2,k}$, ..., $TR_{i,4,k}$, $TR_{i+1,1,k}$,

For each LOS calculate the annual loss rate for prediction

($P_{i,k}$) using single exponential smoothing. The procedure is to iterate through four quarters of data which results in a single value for each year. The superscript (j), $j = 1, 2, 3, 4$, is used to indicate the intermediate steps of

the procedure but need not be carried once the annual predicted loss rate has been calculated.

$$P_{i,k}^{(1)} = \alpha TR_{i,1,k} + (1.0 - \alpha) P_{i-1,k}$$

Where

$P_{i-1,k}$ is the final result from the previous year.

$$P_{i,k}^{(j)} = \alpha TR_{i,j,k} + (1.0 - \alpha) P_{i,k}^{(j-1)} \text{ for all } j = 2, 3, 4$$

$$P_{i,k} = P_{i,k}^{(4)}$$

Note that only one loss rate prediction is made for each year. This means that seasonal variations in the loss rate are not taken into account. Another approach would be to predict a loss rate separately for each quarter, the trade-off being that this procedure would require more years of data. This raises the question of whether using older data which takes into account seasonal variations would result in a better prediction than not using this older data but ignoring the seasonal variation. This is an area left for further study.

The prediction is now made for each quarter.

Let $T_{i,j,k}$ be the prediction for year i (first future year) quarter j and LOS k

$$T_{i,j,k} = A_{i-1,j,k-1}(1.0 - P_{i,k}) \quad \text{for all } j = 1,2,3,4$$

As with the NAPPE model, this gives predictions for LOS cells 2-31. The calculation for cell 1 was done in a manner similar to the existing NAPPE model. The average proportion of the total force in cell 1 was calculated for the four years of data. For each quarter the following calculations were made:

Let ClAV be the number which would be required in cell 1 calculated by taking the above proportion of the total force requirement for the quarter being predicted.

$$\text{Let ClP} = \text{Req} - \sum_{k=2}^{31} T_{i,j,k} \quad \begin{array}{l} \text{total required minus} \\ \text{the sum of cells 2-31} \end{array}$$

Half of the difference between these two values was then allocated among cells 2-31 on the basis of the number projected for that cell.

For each cell calculate

$$ADJ_{i,j,k} = \frac{(ClAV - ClP)}{2} \frac{T_{i,j,k}}{\sum_{k=2}^{31} T_{i,j,k}}$$

$$T'_{i,j,k} = T_{i,j,k} + ADJ_{i,j,k}$$

The value for cell 1 is then the difference between the total required and the sum of the new predictions for cells 2-31.

2. Results of SMOTHY

The model, as described above, was then run to obtain one year predictions for the last ten years. The following table is a comparison of these results with outputs from NAPPE for the same time periods.

TABLE II

COMPARISON OF NAPPE AND SMOTHY PREDICTIONS
(costs are in millions of dollars)

Year	Actual Cost	NAPPE Prediction	% Error	SMOTHY Prediction	% Error
1965	1,362.1	1,360.0	.151	1,358.1	.209
1966	1,582.4	1,591.6	.501	1,582.0	.028
1967	1,720.7	1,736.5	.918	1,737.9	.997
1968	1,832.6	1,838.8	.337	1,836.9	.237
1969	2,009.2	2,016.7	.372	2,015.7	.322
1970	2,280.8	2,286.8	.262	2,280.9	.001
1971	2,264.8	2,258.2	.290	2,256.2	.379
1972	2,496.7	2,480.7	.643	2,483.6	.524
1973	2,683.6	2,678.1	.205	2,678.3	.196
1974	2,777.4	2,773.6	.138	2,772.1	.190
		Mean	.3817		.3083
		Mean 2	.3221		.2318

The value given in the table as mean is the mean of the absolute errors and the value called mean 2 is the same except the outlier (1967) is left out of the calculation. Leaving this value out is not unreasonable when the events of 1967 are taken into consideration. It was during this year that the structure of the force saw tremendous change due to the Viet Nam buildup. Any model based on past data cannot predict the future when major policy decisions make that data inappropriate. This is a point where the analyst using the model must use reason when looking at its output, a point to be discussed later.

A close inspection of the preceeding table yields some surprising conclusions. Although the SMOTHY model does not predict uniformly better, it does significantly better for most years. This suggests that the complexities of the NAPPE model are not only unnecessary, but have a negative effect.

C. DOUBLE EXPONENTIAL SMOOTHING

The most important hypothesis tested was that single exponential smoothing is not the appropriate tool for modeling a time series which appears to have trends. As suggested by Brown [1], Goodman [2], and others, higher-order exponential smoothing is a valuable tool for modeling time series with underlying trends. Since the time series under consideration does not show any properties which would indicate anything

beyond a linear trend, only double smoothing was considered in SMOTH2.

1. The SMOTH2 Subroutine

As with the SMOTHY subroutine, the SMOTH2 subroutine makes use of only four years of historical data and only the TOTALNAV inventories. It also only smoothes the loss rates. These loss rates ($TR_{i,j,k}$) are calculated exactly the same as in SMOTHY, and the loss rate to be used for prediction ($D_{i,k}$) is calculated as described by Brown [1]. The single smoothed portion ($P_{i,k}$) is calculated exactly as before. The double smoothing term is calculated by simply smoothing the single smoothed value. The superscript notation is again used for the four iterations exactly as used to calculate $P_{i,k}$.

$$S_{i,k}^{(1)} = \alpha P_{i,k}^{(1)} + (1.0 - \alpha) S_{i-1,k}$$

$$S_{i,k}^{(j)} = \alpha P_{i,k}^{(j)} + (1.0 - \alpha) S_{i,k}^{(j-1)} \quad \text{for all } j = 2, 3, 4$$

$$S_{i,k} = S_{i,k}^{(4)}$$

These two values are then combined to pick up the linear trend and result in;

$$D_{i,k} = 2 P_{i,k} - S_{i,k} + \frac{\alpha}{1.0 - \alpha} (P_{i,k} - S_{i,k})$$

Note that again only one value of the smoothed loss rate is calculated for each year and the same procedural question was left unanswered.

The prediction is then made for each quarter into the future exactly as in the SMOTHY subroutine

$$T_{i,j,k} = A_{i-1,j,k-1}(1.0 - D_{i,k})$$

As with the previously discussed models, this gives a prediction for LOS cells 2-31. The final adjustments used in SMOTH2 are exactly the same as used in SMOTHY.

2. Results of SMOTH2

As in the experiment with SMOTHY, it was obvious that the most recent data should be most heavily weighted. Therefore, an initial value of 0.4 was used for alpha. Since the impact of trend was the most important consideration of the research, other values of alpha were also tried. The following table is the result of these tests. In order to put the results in a form for analysis, the actual dollar values were not tabled but only the percentage errors. The elements of the table are the percentage error from the actual total cost of the force for NAPPE, NAPPE with the SMOTHY subroutine (these values are the same as Table II), and for NAPPE with the SMOTH2 subroutine using values of alpha of 0.2, 0.3, and 0.4.

TABLE III
COMPARISON OF ABSOLUTE PERCENTAGE ERRORS
OF NAPPE, SMOTHY, AND SMOTH2

Year	NAPPE	SMOTHY	$\alpha=.2$	SMOTH2 $\alpha=.3$	$\alpha=.4$
1965	.151	.209	.242	.223	.215
1966	.501	.028	.015	.027	.018
1967	.918	.997	1.098	1.158	1.170
1968	.337	.237	.196	.098	.273
1969	.372	.322	.133	.033	.054
1970	.262	.001	.205	.076	.041
1971	.290	.379	.443	.423	.473
1972	.643	.524	.510	.527	.543
1973	.205	.196	.171	.164	.147
1974	.138	.190	.132	.068	.059
Mean	.3817	.3083	.3145	.2797	.2993
Mean 2	.3221	.2318	.2274	.1821	.2026
Mean 3	.2801	.1690	.1563	.0984	.1153

The values of mean and mean 2 have the same definition as in the preceeding table. The value mean 3 was calculated leaving out the values for 1967, 1971, and 1972. The reason for making this calculation will be discussed in detail later.

The first overview of the table would result in a conclusion that SMOTH2, with an alpha value of 0.3 seems to be a somewhat better model on the basis of the mean alone.

However, the mean is not the only significant feature. The most important observation is that SMOTH2 predicts extremely well for all years except 1967, 1971, and 1972. The reason for the poor prediction in 1967 has already been discussed. The reason for the poor prediction for 1971 and 1972 can be explained in the same manner except that the force structure was moving in the opposite direction. That is, these were the years of major policy changes resulting from the end of the Viet Nam War and the shrinking of the force.

Because of the linear trend, which is a part of SMOTH2, it must be expected to do poorly when the direction of the trend changes. This means that SMOTH2 should have more difficulty "turning the corner" when there are major policy changes. This does not mean that it is a poor model, but rather, some decision rule is required of the user when this occurs.

D. THE MATRIX GENERATION PROCEDURE

Although incomplete and inconclusive, a study of the Mosteller method, as used in this model, had some interesting results. It was found that, in general, the Mosteller procedure is extremely sensitive to the base matrix. It was found that changing the value in one cell of the base matrix resulted in changing the values in virtually every cell of the output matrix. There was no consistency found in these changes. The surprising result was that the changes in the output matrix were sometimes greater than the initial changes

in the base matrix. For example, changing one cell value in the base matrix by less than 5% could result in changes in the output matrix of greater than 5% in some cells.

This is extremely significant when considering its use in this model. The base matrix is calculated as a simple average of the last twelve quarterly inventories. This means that seasonal variations in the force structure are not taken into account and implies that a better base matrix may be possible.

A hypothesis was made that a better base matrix could be calculated by computing some very rough transition rates from one cell to another and these rates applied to the one year previous inventory. Experiments with this hypothesis showed some very promising results but were inconclusive. Continued study in this area may be of considerable value.

IV. CONCLUSIONS

The NAPPE model, in its present form of single exponential smoothing, does not appear to be the appropriate model. Single exponential smoothing has as an assumption that the time series is basically constant in time and the difference from the mean is caused by some random noise. This does not appear to be the case with LOS populations or with transition rates.

A recommended change to the model is to remove the complex SMOOTH and ADJSMO subroutines and replace them with the SMOTH2 subroutine, using an alpha of 0.3. It should be made clear to any intended user, however, that substantial changes in the enlisted force management would not be reflected in the prediction. The modelling approach should, in fact, be completely revised so that changes of this magnitude can be accounted for. Since pay grade totals are used to drive the force structure, the model is aware of impending changes in direction. This information is not currently being used in loss prediction by NAPPE.

In addition, the base matrix used in the Mosteller procedure could be estimated more carefully. Based on these preliminary experiments, this could result in a much better estimate of force structure, and hence a more accurate budget prediction. The determination of LOS cell 1 population remains somewhat ad hoc, as does the choice of a smoothing

constant of 0.3. While this study has demonstrated that a simpler approach to the prediction can be successfully taken, all other alternatives have not been thoroughly investigated.

APPENDIX A

This appendix is a rough documentation of the SMOOTH subroutine in the current NAPPE model.

Let

$A_{i,j,k}$ be the actual population for year i
quarter j and LOS cell k

For each LOS $k = 1, 2, \dots, 30$, calculate the transition rate
for each year and quarter

$$TR_{i,j} = \frac{A_{i,j,k} - A_{i+1,j,k+1}}{A_{i,j,k}}$$

For each year $i = 1, 2, \dots, \text{NYR}-1$ (NYR = last year of
historical data) and each $\alpha = .05, .10, \dots, .95$, calculate:

$$P_{5}^1 = TR_{1,1}$$

Let $n = 4i + j$ $j = 1, 2, 3, 4$

$$P_{n+1}^1 = \alpha TR_{i,j} + (1.0 - \alpha) P_n^1$$

For each year then the predicted transition rate is

$$PRED_{i+2,\alpha} = P_{4i+1}^1$$

The relative error in this prediction is then the sum of the differences between the predicted and actual transition rates for the year

$$ER_{i+2,\alpha} = \sum_{j=1}^4 \left| \frac{TR_{i+1,j} - P_{4i+1}}{1 - TR_{i+1,j}} \right|$$

Go to the INLINE subroutine to choose the best α .

INLINE Subroutine

For each year $i = 1, 2, \dots$, NYR find the best α for predicting the following year.

For each $\alpha = .05, .10, \dots, .95$ calculate

$$EMIN_1 = ER_{i,\alpha}$$

$$EMIN_2 = \sum_{\ell=1}^i ER_{\ell,\alpha}$$

$$EMIN_3 = \sum_{\ell=1}^i (ER_{\ell,\alpha})^2$$

Select the α which gives the minimum value of $EMIN_1$, $EMIN_2$, $EMIN_3$ and call them α_1^* , α_2^* , α_3^* .

Calculate

$$SE_1 = \sum_{\ell=1}^i (ER_{\ell,\alpha_1^*})^2$$

$$SE_2 = \sum_{\ell=1}^i (ER_{1,\alpha_2^*})^2$$

Within the summation here, each α^* is the one which was selected for the given year

$$SE_3 = \sum_{\ell=1}^i (ER_{1,\alpha_2^*})^2$$

Select the minimum of these three values and the most recent α^* for that method is the α to be used to predict the following year. Call this value α_{i+1}

For each year $i = 1, 2, \dots, \text{NYR}$ and each quarter of the year, make the prediction based on the transition rate:

$$T_{i,j,k} = A_{i-1,j,k} (1.0 - \text{PRED}_{i,\alpha_i})$$

Now make a similar prediction based on population in each cell.

For each year $i = 2, 3, \dots, \text{NYR}$ and each $\alpha = .05, .10, \dots, .95$

$$P_{8}^1 = A_{1,4,k+1}$$

Let $n = 4i+j$ $j = 1, 2, 3, 4$

$$P_{n+1}^1 = \alpha A_{i,j,k+1} + (1.0 - \alpha) P_n^1$$

$$\text{PRED}_{i+1,\alpha} = P_{4i+1}^1$$

The relative error in this prediction is then

$$\text{ER}_{i+1,\alpha} = \sum_{j=1}^4 \left| \frac{A_{i+1,j,k+1} - P_{4i+1}^1}{A_{i+1,j,k+1}} \right|$$

Go to the INLINE subroutine to choose the best α . For each year $i = 2, 3, \dots, \text{NYR}$ and each quarter (the same value is predicted for all four quarters of a given year), make the prediction based on cell populations

$$P_{i,k} = \text{PRED}_{i,\alpha_i}$$

APPENDIX B

This appendix is a rough documentation of the ADJSMO subroutine in the current NAPPE model.

Define the five methods or techniques used in the subroutine:

1. $P(1)_{i,j} = T_{i,j,k}$ this is the predicted value for year i , quarter j calculated as the transition rate based prediction from SMOOTH using the TOTALNAV data.
2. $P(2)_{i,j} = T_{i,j,k}$ this is the sum of the predictions from SMOOTH made as above using USN and USNR data.
3. $P(3)_{i,j} = P_{i,k}$ this is the predicted value for year i , quarter j calculated as the population based prediction from SMOOTH using the TOTALNAV data.
4. $P(4)_{i,j} = P_{i,k}$ this is the sum of the predictions from SMOOTH made as above using USN and USNR data.

$$5. \quad P(5)_{i,j} = \text{BWT} \frac{P(1) + P(2)}{2} + (1.0 - \text{BWT}) \frac{P(3) + P(4)}{2}$$

this is a weighted average of
1-4 where the value of BWT is
the value which would have
predicted best for the previous
year.

For each LOS $k = 2, 3, \dots, 31$

For each year $i = 1, 2, \dots, \text{NYR}$ calculate the cumulative
square error for each method $I = 1, 2, 3, 4, 5$.

$$\text{TEP}(I)_i = \sum_{n=1}^i \sum_{j=1}^4 \left(\frac{A_{i,j,k} - P(I)_{n,j}}{A_{i,j,k}} \right)^2$$

For each year calculate the cumulative square error
for all values of $\text{BWT} = .45, .50, \dots, .95$

$$\text{ET2}(\text{BWT})_i = \sum_{n=1}^i \sum_{j=1}^4 \left(\frac{A_{i,j,k} - X_{i,j}(\text{BWT})}{A_{i,j,k}} \right)^2$$

where

$$X_{i,j}(\text{BWT}) = \text{BWT} \frac{P(1)_{i,j} + P(2)_{i,j}}{2} + (1.0 - \text{BWT}) \frac{P(3)_{i,j} + P(4)_{i,j}}{2}$$

Based on TEP select the best method, I, and based on ET2 select the best BWT which will then be used to calculate P(5) for the following year.

Calculate the cumulative absolute error for the entire period using the method which was selected as best for each year.

$$TER_k = \sum_{L=1}^{NYR} \sum_{S=1}^4 \left| \frac{A_{i,j,k} - P(I_{n-1})_{i,j}}{A_{i,j,k}} \right| \quad \text{where } I_{n-1} \text{ was the best method for that year.}$$

Make the initial prediction for the first future year

$$A_{NYR+1,j,k+1} + P(I_{NYR})_{NYR+1,j} \quad \text{where } I_{NYR} \text{ is the method selected best on the last year of data.}$$

Let $IPG_{i,j}$ be the total force required

Calculate the average proportion of the total

force in cell 1 for all years of data, call it PAV

For each quarter to be predicted, calculate

$$ClAV = (PAV) (IPG)$$

$$ClP = IPG - \sum_{k=2}^{31} A_{NYR+1,j,k}$$

These two values are the possible predictions for LOS 1. ClAV is based on the average proportion of the force in cell 1, while ClP is simply the difference between the total required force and the predictions for cells 2-31. Let

$$ClADJ = \frac{ClAV + ClP}{2}$$

There is a test in the model to ensure that this average is between the values which would have been calculated using the largest and smallest proportions of the total population in cell 1 over the entire data base.

Take the difference between ClADJ and ClP and allocate it among cells 2-31 according to the total error which was calculated for predicting that cell using the best method.

$$A_{i,j,k}^* = A_{i,j,k} + \frac{(ClP - ClADJ) TER_{k-1} A_{i,j,k}}{\sum_{k=2}^{31} TER_{k-1} A_{i,j,k}}$$

Adjust cell 1 from these values

$$A_{i,j,l}^* = IPG_{i,j} - \sum_{k=1}^{31} A_{i,j,k}^*$$

The same basic procedure is used for predicting additional future years (up to 10) and the value of continuing the discussion is questionable.


```

// EXEC FORTCLG,REGION.GO=150K
//FORT.SYSRINT DD DUMMY
//FORT.SYSIN DD *
COMMON NYR,NYRM1,NYRP1,IST,NQTBP,NCTS,ILJ(10),NUM(10,32),
$ NUM(10,32),NUMA(10,32),IPG(40,10),RATE(10,32),COST(10,4),
$ APOP(32,4,32),PPOP1(30,30),PPOP2(30,30),TPOP1(30,4,30),
$ TPOP2(30,4,30),
$ IRED,ILM,ILY,
$ IRT,IPUNCH,LHAVE(40)
LOGICAL IRT,IPUNCH,LHAVE
DIMENSION XNLOS(10)
INTEGER APOP
DO 2 IYR=1,30
DO 2 LOS=1,30
PPOP2(IYR,LOS)=0.0
DO 2 IQTR=1,4
TPOP2(IYR,IQTR,LOS)=0.0
2 CONTINUE
C CALL SUBROUTINE TO READ & EDIT CONTROL CARDS
C
IRED=0
CALL RDCONT
IF (IRT) WRITE (6,6000)
6000 FORMAT (' CONTROL CARDS READ')
C
IST=3
CALL RDINV (ILM,ILY)
IF (IRT) WRITE (6,6002) IST
6002 FORMAT (' INVENTORIES READ ',I2)
C CALL SMOTH2
C FOR EACH QTR FLIP-FLOP MATRIX & CSOT FORCE
C
IMO=ILM
IYR=ILY
DO 20 NQ=1,NQTBP
IMO=IMO+3
IF (IMO.LT.13) GO TO 15
IMO=3
IYR=IYR+1
CONTINUE
15 CALL PNGPNG (IPG,APOP,NQ,NYR,NUMA,NUM)
WRITE(7,7000) IMO,IYR,NQ,(J,(NUM(I,J)),I=1,10),J=1,32)
7000 FORMAT ('1,35X,PREDICTION GF ALL NAVY ENLISTED INVENTORY FOR',I300000690
$ ' ',I2/45X,'PROJECTION LENGTH =',I3,' QUARTERS',//
$ 'X',LOS,12X,'E-1',8X,'E-2',8X,'E-3',8X,'E-4',8X,'E-5',8X,'E-6',
$ 'E-7',8X,'E-8',8X,'E-9',6X,'TOTAL',/32(/4X,I2,4X,10I11))
IF (IPUNCH) WRITE(4,5000) (ILM,ILY,IMO,IYR,(NUM(I,J),I=1,10),J,

```

```

000000010
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000000050
000000060
000000070
000000080
000000090
000000100
000000110
000000120
000000130
000000140
000000150
000000160
000000170
000000180
000000190
000000200
000000210
000000220
000000260

000000280
000000290
000000300

000000570
000000580
000000590
000000600
000000610
000000620
000000630
000000640
000000650
000000660
000000670
000000680
000000690
000000700
000000710
000000720
000000730

```



```

5000 $ J=1,32)
      FORMAT (4I2,10I7,12)
      WRITE (7,7001)
      DO 30 MN=1,10
      X=0.0
      DO 31 LOS=1,31
      X = X + (LOS-.5)*NUM(MN,LOS)
      31 M = NUM(MN,32)
      IF (M.EQ.0) XMNLOS(MN)=0.0
      IF (M.NE.0) XMNLOS(MN) = X/M
      30 WRITE (7,4000) (XMNLOS(MN),MN=1,10)
      4000 FORMAT(10,8X,'E-1',9X,'E-2',9X,'E-3',9X,'E-4',9X,'E-5',9X,'E-6',
      $ 9X,'E-7',9X,'E-8',9X,'E-9',7X,'TOTAL',/10F12.3)
      7001 FORMAT(10,20X,'MEAN LOS DISTRIBUTION FOR THE QUARTER')
      CALL COSTPG (IMO,IYR)
      CONTINUE
      RETURN
      END
20

```

```

C
C
C SUBROUTINE RDCONT

```

```

COMMON NYR,NYRMI,NYRPI,IST,NQTBP,NCTS,ILJ(10),NUM(10,32),
$ NUMC(10,32),NUMA(10,32),IPG(40,10),RATE(10,32),COST(10,4),
$ APCP(32,4,32),PPOP1(30,30),PPOP2(30,30),TPOP1(30,4,30),
& TPOP2(30,4,30),
$ IRED,ILM,ILY,
$ IRT,IPUNCH,LHAVE(40)
LOGICAL IRT,IPUNCH,LHAVE
INTEGER APCP

```

```

C
C
C DIMENSION HEAD(19)
C
C WRITE HEADER
C

```

```

DO 10 I=1,21
  READ (11,1100) (HEAD(J),J=1,19)
  10 WRITE (7,7000) (HEAD(J),J=1,19)
  1100 FORMAT(19A4)
  7000 FORMAT (25X,19A4)

```

```

C READ CONTROL INFO
C
C READ (5,5000) ILM,ILY,NQTBP,IRTT,IPNCH
5000 FORMAT (3I2,72X,2I1)
C
C IRT=.FALSE.

```

```

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0001020
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0001040
0001050
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0001080
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0001100
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0001150
0001160
0001170
0001180
0001190

```



```

IF (IRTT.EQ.1) IRT=.TRUE.
IPUNCH=.FALSE.
IF (IPNCH.EQ.1) IPUNCH=.TRUE.
IF (NQTBP.LE.40) GO TO 5
WRITE (6,6000) NQTBP
FORMAT (2X,I2,' QUARTERS REQUESTED; ONLY 40 CAN BE PROJECTED')
6000 NQTBP = 40
C
5 WRITE (7,7002) NQTBP,ILM,ILY,IRTT,IPNCH
C
7002 FORMAT (5(/),20X,'NOTES: '//26X,'THE',I3,' QUARTERLY PROJECTIONS MA
$DE ARE BASED ON HISTORICAL DATA THRU',I3,'/19',I2,18X,211,725X,
$USER INPUT PAY GRADE TOTALS USED IN PROJECTIONS ARE:',/20X,'FOR',
$5X,'E-1',5X,'E-2',5X,'E-3',5X,'E-4',5X,'E-5',5X,'E-6',5X,'E-7',5X,
$'E-8',5X,'E-9',3X,'TOTAL')
C
DO 30 I=1,NQTBP
LHAVE(I)=.FALSE.
30
C
NLKQ=ILM/3+ILY*4
C
NYR=(NLKQ-228)/4
NCTS=(NLKQ-228-4*NYR)*32
IF (NYR.LE.20) GO TO 85
NYR=NYR-1
NCTS=NCTS+128
GO TO 80
80
85 NYRMI=NYR-1
NYRPI=NYR+1
C
C
40 READ (5,5001,END=60) IMO,IYR,ILJ
5001 FORMAT (2X,2I2,2X,10I7)
NQPG=IMO/3+IYR*4-NLKQ
IF ((NQPG.LT.1).OR.(NQPG.GT.40)) GO TO 45
IF (LHAVE(NQPG)) GO TO 45
DO 50 I=1,10
IPG(NQPG,I)=ILJ(I)
50 WRITE (7,7004) IMO,IYR,ILJ
7004 FORMAT (19X,I2,'/ ',I2,10I8)
LHAVE(NQPG)=.TRUE.
GO TO 40
C
45 WRITE (6,6001) IMO,IYR,ILJ
6001 FORMAT (' PAY GRADE CARD:',2I3,10I8,2X,'NOT USED')
GO TO 40
C
C CHECK FOR SUFFICIENT PAY GRADE TOTALS

```

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00001500
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00001580
00001590
00001600
00001610
00001620
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00001650
00001660
00001670


```

C 60      DO 65 I=1,NQTBP
          IF (.NOT.LHAVE(I)) GO TO 70
          CONTINUE
          RETURN
C 70      IMO=ILM-2
          IF (ILM.GT.6) IMO=(18-ILM)/3
          IF (.NOT.LHAVE(IMO)) GO TO 90
          IF (IMO.GE.NQTBP) RETURN
          IMO=IMO+4
          GO TO 75
C 90      WRITE (6,6002)
6002      FORMAT (' INSUFFICIENT PAY GRADE TOTALS INPUT')
          RETURN
          END
C

```

```

SUBROUTINE ROINV
COMMON NYR,NYRMI,NYRPI,IST,NQTBP,NCTS,ILJ(10),NUM(10,32),
$ NUMO(10,32),NUMA(10,32),IPG(40,10),RATE(10,32),COST(10,4),
$ APOP(32,4,32),PPOP1(30,30),PPOP2(30,30),IPOP1(30,4,30),
$ TPDP2(30,4,30),
$ IRED,ILM,ILY,
$ IRT,IPUNCH,LHAVE(40)
LOGICAL APOP
INTEGER FJUNE
LOGICAL FJUNE
IF (IST.NE.3) GO TO 4
FJUNE=.FALSE.
NYRM2=NYR-2
DO 2 I=1,10
DO 2 J=1,32
NUMA(I,J)=0
IF (NCTS.EQ.0) GO TO 6
DO 5 I=1,NCTS
READ (IST,1000)
FORMAT (80X)
CONTINUE
DO 50 IYR=1,NYR
DO 50 IQTR=1,4
READ (IST,1005,END=100) IMO,JYR,((NUM(I,J),I=1,10),J=1,32)
FORMAT (2X,2I2,2X,10I7,31(/8X,10I7))
DO 10 J=1,32
APOP(IYR,IQTR,J)=NUM(10,J)
IF (IST.NE.3) GO TO 50
IF (IYR.LT.NYRM2) GO TO 50

```

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00001680
00001690
00001700
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0002100
0002110
0002120
0002130

```



```

DO 20 I=1,10
DO 20 J=1,32
    NUMA(I,J)=NUMA(I,J)+NUM(I,J)
IF (IYR.NE.NYR) GO TO 50
IF (FJUNE) CALL COSTPG(IMO,JYR)
IF (IMO.NE.6) GO TO 50
FJUNE=.TRUE.
DO 40 I=1,10
DO 40 J=1,32
    ILJ(I)=NUM(I,32)
DO 40 J=1,32
    NUMO(I,J)=NUM(I,J)

CONTINUE
REWIND IST
RETURN
WRITE(6,6000) IST
FORMAT(' INSUFFICIENT DATA ON FILE',I2)
END

C C C C C
SUBROUTINE SMOH2
COMMON NYR,NYRM1,NYRP1,IST,NQIBP,NCTS,ILJ(10),NUM(10,32),
$ $ $ $ $
$ $ $ $ $
$ $ $ $ $
$ $ $ $ $
$ $ $ $ $
DIMENSION TR(20,4),PREDD(20),TPOP(32,4,32)
INTEGER APOD
NQ=0
LNY=NYR
INY=NYR-4
JNY=NYRM1
KNY=NYRP1
CCONTINUE
DO 200 LOS=1,30
LOSPI=LOS+1
DO 10 IYR=INY,JNY
DO 10 IQTR=1,4
SPOP=APUP(IYR,IQTR,LOS)
IF(LOS.EQ.30) SPOP=SPOP+APOP(IYR,IQTR,31)/SPOP
10 TR(IYR,IQTR)=(SPOP-APOP(IYR+1,IQTR,LOSPI))/SPOP
ALF=.4
S1=TR(INY,1)
S2=S1
ISQ=2

```



```

DO 30 IYR=INY,JNY
DO 20 IQTR=ISQ,4
S1=ALF*TR(IYR,IQTR) + (1.0-ALF)*S1
S2=ALF*S1 + (1.0-ALF)*S2
ISQ=1
CONTINUE
20 CONTINUE
30 AHAT=2.0*S1 - S2
BHAT=(ALF/(1.0-ALF))*(S1-S2)
PRED(IYR+2)=AHAT+BHAT
DO 75 IQTR=1,4
75 TPOP(KNY,IQTR,LOSP1)=APOP(LNY,IQTR,LOS)*(1.0-PRED(KNY))
200 CONTINUE
PAV=0.0
DO 120 IYR=INY,LNY
DO 120 IQ=1,4
R=APOP(IYR,IQ,1)
R=R/APOP(IYR,IQ,32)
120 PAV=PAV+R
PAV=PAV/20.
IYR=KNY
DO 160 IQ=1,4
NQ=NQ+1
TP=0.0
CIAV=PAV*IPG(NQ,10)
CIP=IPG(NQ,10)
DO 130 LOS=2,31
CIP=CIP-TPCP(IYR,IQ,LOS)
130 TP=TP+TPCP(IYR,IQ,LOS)
C1=.5*(CIP+CIAV)
APOP(IYR,IQ,1)=IPG(NQ,10)
DO 135 LOS=2,31
ADJ=(CIP-C1)*TPCP(IYR,IQ,LOS)/TP
APOP(IYR,IQ,LOS)=APOP(IYR,IQ,1) + ADJ
135 APOP(IYR,IQ,1)=APOP(IYR,IQ,LOS)
160 CONTINUE
IF(NQ.EQ.NQTBP) GO TO 201
LNY=LNY+1
INY=INY+1
JNY=JNY+1
KNY=KNY+1
GO TO 202
201 RETURN
END

```

C

C SUBROUTINE PNGPNG (IPG,APOP,NQ,NYR,NUM,NUM2)

C SUBROUTINE PNGPNG FINDS INVENTORY NUM2 THAT HAS SIDE TOTALS ISIDE 00005020
00005030


```

C AND FOOT TOTALS IFOOT THAT HAS APPROXIMATELY THE SAME DISTRIBUTION
C AS NUM.
DIMENSION ISIDE(32),IFOOT(10),NUM(10,32),NUM2(10,32),XPRES(32,10)
DIMENSION ID(31),JD(9)
INTEGER APOP
DIMENSION IPG(40,10),APOP(32,4,32)
DO 1 I=1,10
  IFOOT(I)=IPG(NQ,I)
  J=(NQ+3)/4
  K=NQ-J*4+4
  J=J+NYR
  DO 2 I=1,31
    ISIDE(I)=APOP(J,K,I)
  DO 10 J=1,9
    GOT=IFOOT(J)
    WANT=NUM(J,32)
    IF (WANT.LE.0.) GO TO 5
    WANT=GOT/WANT
  CONTINUE
  DO 10 I=1,31
    GOT=NUM(J,I)
    XPRES(I,J)=GOT*WANT
  DO 90 IC=1,200
    SR=0.
    TE=0.
    DO 50 I=1,31
      GOT=0.
      DO 30 J=1,9
        X=XPRES(I,J)-NUM(J,I)
        TE=TE+ABS(X)
        GOT=GOT+XPRES(I,J)
      WANT=ISIDE(I)
      R=ABS(WANT-GOT)
      IF (R.LE.1.) GO TO 50
      SR=SR+R
      IF (GOT.LE.0.) GO TO 50
      R=WANT/GOT
      DO 40 J=1,9
        XPRES(I,J)=XPRES(I,J)*R
      CONTINUE
    IF (SR.LE.1.) GO TO 100
    SR=0.
    TE=0.
    DO 80 J=1,9
      GOT=0.
      DO 60 I=1,31
        X=XPRES(I,J)-NUM(J,I)
        TE=TE+ABS(X)

```

```

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00005690
00005700
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00005720
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00005740
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00005770
00005780
00005790

```

60      GOT=GOT+XPRE(I,J)
      WANT=IFOOT(J)
      R=ABS(WANT-GOT)
      IF (R.LE.1.) GO TO 80
      SR=SR+R
      IF (GOT.LE.0.) GO TO 80
      R=WANT/GOT
      DO 70 I=1,31
        XPRE(I,J)=XPRE(I,J)*R
      CONTINUE
1005    FORMAT (8X,I4,2F15.1)
      IF (SR.LE.1.) GO TO 100
90      CONTINUE
1000    WRITE (6,1000) ERROR : PING PONG DOES NOT CONVERGE.'')
100    FORMAT (110 I=1,31
      NUM2(10,I)=0
      DO 110 J=1,9
        NUM2(J,I)=XPRE(I,J)+.5
        NUM2(10,I)=NUM2(10,I)+NUM2(J,I)
110    DO 120 J=1,10
        NUM2(J,32)=0
        NUM2(J,J,32)=0
      DO 120 I=1,31
        NUM2(J,32)=NUM2(J,32)+NUM2(J,I)
120    RETURN
      END

```

C
C
C

00005800
00005810
00005820
00005830
00005840
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00005870
00005880
00005890
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00005970
00005980
00005990

```

SUBROUTINE COSTPG (IMO,IYR)
COMMON NYR,NYRMI,NYRPI,IST,NQIBP,NCTS,ILJ(10),NUM(10,32),
$ NUM(10,32),NUMA(10,32),IPG(40,10),RATE(10,32),COST(10,4),
$ APQP(32,4,32),PPOP1(30,30),PPOP2(30,30),TPOP1(30,4,30),
$ TPOP2(30,4,30),
$ IRED,ILM,ILY,
$ IRT,IPUNCH,LHAVE(40)
LOGICAL IRT,IPUNCH,LHAVE
INTEGER APQP
LOGICAL LNR
LOGICAL LNR
DIMENSION RATE2(10)
LNR=.TRUE.
J=IMO/3
DO 5 I=1,10
  COST(I,J)=0.0
  PATE2(I)=0.0
5 CONTINUE
IQED=IYR*360+IMO*30+30
10 IF (IQED.LE.IRED) GO TO 20

```



```

P=IQED-IRED
P=P/90.0
READ (8,8000,END=500) IRM,IRD,IRY
IRED=IRY*360+IRM*30+IRD
IF (P.LT.1) GO TO 25
READ (8,8005,END=500) ((RATE(I,K),I=1,10),K=1,31)
GO TO 10
LNR=.FALSE.
P=0.0
PM1=1.0-P
DO 30 I=1,31
  IF (LNR) READ (8,8005,END=500) RATE2
  DO 30 K=1,9
    COST(K,J)=COST(K,J)+(NUMO(K,I)+NUM(K,I))*(RATE(K,I)*PM1+
$    RATE2(K)*P)*1.5
    IF (LNR) RATE(K,I)=RATE2(K)
  CONTINUE
  A=0.0
  DO 40 I=1,9
    A=A+COST(I,J)
  DO 40 K=1,31
    NUMO(I,K)=NUM(I,K)
  CONTINUE
  COST(10,J)=A
  WRITE (7,4005) IMO,IYR
  WRITE (7,4000) (COST(I,J),I=1,10)
  IF (IPUNCH) WRITE (4,7000) (COST(I,J),I=1,10)
  IF (IMG.NE.6) RETURN
  DO 50 J=1,10
    A=0.0
    DO 45 I=1,4
      A=A+COST(J,I)
      RATE2(J)=A
    WRITE (7,4010) IYR
  FORMAT (5F12.0)
  IF (IPUNCH) WRITE (4,7000) RATE2
  WRITE (7,4000) RATE2
  FORMAT (10,8X,'E-1',9X,'E-2',9X,'E-3',9X,'E-4',9X,'E-5',9X,'E-6',
$ 9X,'E-7',9X,'E-8',9X,'E-9',7X,'TOTAL',/10F12.0)
  FORMAT (10,20X,'PROJECTED ENLISTED BASIC PAY FOR QUARTER ENDING',
$ 13,/,12)
  FORMAT (10,20X,'PROJECTED ENLISTED BASIC PAY FOR FISCAL YEAR 19',
$ 12)
  RETURN
8000 FORMAT (3I2)
8005 FORMAT (8X,10F7.2)
500 WRITE (6,6000)
6000 FORMAT (:,$$ ERROR')

```

```

00005990
00006000
00006010
00006020
00006030
00006040
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00006070
00006080
00006090
00006100
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00006360
00006370
00006380
00006390
00006400
00006410
00006420
00006430
00006440
00006450
00006460

```



```

RETURN
END
SUBROUTINE PREDPG
COMMON NYR,NYRML,NYRPI,IST,NQIBP,NCTS,ILJ(10),NUM(10,32),
$ NUM(10,32),NUMA(10,32),IPG(40,10),RATE(10,32),COST(10,4),
$ APOP(32,4,32),PPOP1(30,30),PPOP2(30,4,30),
$ TPOP2(30,4,30),
$ IRED,ILM,ILY,
$ IRT,IPUNCH,LHAVE(40)
LOGICAL IRT,IPUNCH,LHAVE
INTEGER APOP
DIMENSION ST(3,9),PRE(3,10)
DATA ST/1,1248,.8539,1.0040,1.0146,1.0555,1.0426,1.0454,.9768,
11.0376,.9380,.9848,.9557,.9361,.9712,.9389,.9619,.9856,
2.9687,1.0086,.9355,.9831,.9661,1.0284,1.0191,.9705,1.0142,1.0054/
NJ=ILM/3
IF (NJ.NE.1) NJ=6-NJ
DO 10 I=1,NQIBP
5 IF (.NOT.LHAVE(I)) GO TO 20
CONTINUE
RETURN
CONTINUE
IF (.NOT.LHAVE(NJ)) GO TO 100
20 IF (I.LT.NJ) GO TO 40
GO TO 66
PRE(1,10)=0.
40 PRE(2,10)=0.
PRE(3,10)=0.
DO 60 K=1,9
PRE(2,K)=(ILJ(K)+IPG(NJ,K))/2
PRE(1,K)=(ILJ(K)+PRE(2,K))/2
PRE(3,K)=(PRE(2,K)+IPG(NJ,K))/2
DO 50 J=1,3
PRE(J,K)=PRE(J,K)*ST(J,K)
50 PRE(J,10)=PRE(J,10)+PRE(J,K)
60 CONTINUE
DO 65 K=1,3
I=NJ-K
IF (I.LE.0) GO TO 65
IF (LHAVE(I)) GO TO 65
LHAVE(I)=.TRUE.
62 DO 62 J=1,10
65 IPG(I,J)=PRE(4-K,J)+.5
CONTINUE
DO 70 I=1,10
66 ILJ(I)=IPG(NJ,I)
70 NJ=NJ+4

```


100 GO TO 5
1000 WRITE (6,1000)
FORMAT ('\$\$\$\$\$ERROR : JUNE PAY GRADE TOTALS MUST BE INPUT')
RETURN
END

00006950
00006960
00006970
00006980
00006990


```

CALL SMOTHY SMOTHY
SUBROUTINE SMOTHY
COMMON NYR, NYRMI, NYRPI, IST, NQTBP, NCTS, ILJ(10), NUM(10,32),
$ NUMC(10,32), NUMA(10,32), IPG(40,10), RATE(10,32), COST(10,4),
$ APOP(32,4,32), PPOP1(30,30), PPOP2(30,30), TPOP1(30,4,30),
$ TPOP2(30,4,30),
$ IRED, ILM, ILY,
$ IRT, IPUNCH, LHAVE(40)
DIMENSION TR(20,4), PRED(20), TPOP(32,4,32)
INTEGER APOP
NQ=0
LNY=NYR
INY=NYR-4
JNY=NYRMI
KNY=NYRPI
CONTINUE
DO 200 LOS=1,30
  LOSPI=LOS+1
  DO 10 IYR=INY, JNY
    DO 10 IQTR=1,4
      SPOP=APOP(IYR, IQTR, LOS)
      IF(LOS.EQ.30) SPOP=SPOP+APOP(IYR, IQTR, 31)
10  TR(IYR, IQTR)=(SPOP-APOP(IYR+1, IQTR, LOSPI))/SPOP
    ALF=.4
    PI=TR(INY, 1)
    ISQ=2
    DO 30 IYR=INY, JNY
      DO 20 IQTR=ISQ, 4
        PI=ALF*TR(IYR, IQTR)+(1.0-ALF)*PI
      ISQ=1
20  CONTINUE
30  CONTINUE
    PRED(IYR+2)=PI
    DO 75 IQTR=1,4
75  TPOP(KNY, IQTR, LOSPI)=APOP(LNY, IQTR, LOS)*(1.0-PRED(KNY))
200 CONTINUE
    PAV=0.0
    DO 120 IYR=INY, LNY
      DO 120 IQ=1,4
        R=APOP(IYR, IQ, 1)
        R=R/APOP(IYR, IQ, 32)
120  PAV=PAV+R
    PAV=PAV/20.
    IYR=KNY
    DO 160 IQ=1,4
      NQ=NQ+1
      TP=0.0
    CLAV=PAV*IPG(NQ, 10)

```

00


```

CIP=IPG(NQ,10)
DO 130 LOS=2,31
  CIP=CIP-TPOP(IYR,IQ,LOS)
130 TP=TP+TPOP(IYR,IQ,LOS)
  C1=.5*(CIP+C1AV)
  APOP(IYR,IQ,1)=IPG(NQ,10)
DO 135 LOS=2,31
  ADJ=(CIP-C1)*TPCP(IYR,IQ,LOS)/TP
  APOP(IYR,IQ,LOS)=TPOP(IYR,IQ,LOS) + ADJ
135 APOP(IYR,IQ,1)=APOP(IYR,IQ,1) - APOP(IYR,IQ,LOS)
160 CONTINUE
  IF(NQ.EQ.NQTBP) GO TO 201
  LNY=LNY+1
  INY=INY+1
  JNY=JNY+1
  KNY=KNY+1
  GO TO 202
201 RETURN
END

```


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